

Model: Output = $\beta_0 + \beta_1 \cdot \text{Cost} + \epsilon$
 (DV) (IV)

N=90
 K=2 (β_0, β_1)

regress output cost

Source	SS	df	MS
Model	107.123089	1	107.123089
Residual	10.7034329	88	.12162992
Total	117.826522	89	1.32389351

Number of obs = 90
 F(1, 88) = 880.73
 Prob > F = 0.0000
 R-squared = 0.9092
 Adj R-squared = 0.9081
 Root MSE = .34875

output	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
cost	.9691952	.032658	29.68	0.000	.9042942 1.034096
_cons	-14.12819	.4380397	-32.25	0.000	-14.99871 -13.25768

$\frac{SP_{xy}}{SS_{xx}}$

$\bar{y} - b_1 \bar{x}$

tabstat output cost, stat(n mean)

stats	output	cost
N	90	90
mean	-1.174309	13.36561

gen SSxx=(cost-13.36561)^2
 gen SSyy=(output-(-1.174309))^2
 gen SSxy=(cost-13.36561)*(output-(-1.174309))

tabstat cost output SSxx SSyy SPxy, stat(mean sum)

stats	cost	output	SSxx	SSyy	SPxy
mean	13.36561	-1.174309	1.267121	1.309184	1.228088
sum	1202.905	-105.6878	114.0409	117.8265	110.5279

di 110.5279/114.0409
 96919526 → b_1

di -1.174309 - 96919526 * 13.36561
 -14.128195 → b_0

gen yhat=-14.12819+.9691952*cost
 gen ee=(output-yhat)^2
 tabstat ee, stat(sum)

variable	sum
ee	10.70343

di 10.70343/(90-2)
 .12162989

di sqrt(.12162989/114.0409)
 .03265802

di .9691952/.03265802
 29.677096

di sqrt(.12162989*(1/90+13.36561^2/114.0409))
 .43803968

di -14.12819/.43803968
 -32.25322

di 117.826522-10.70343
 107.12309

$K-1$
 $N-K$

MSM
 MSE

$\frac{MSM}{MSE}$

$\frac{SSM}{SST}$

$\sqrt{MSE} = \sqrt{.1216}$

= S_{ϵ}

$MSE = S_{\epsilon}^2$
 $= \hat{\sigma}_{\epsilon}^2$

$\frac{MSE}{SS_{xx}} = \sqrt{\frac{S_{\epsilon}^2}{SS_{xx}}}$

$\sqrt{MSE(\frac{1}{n} + \frac{\bar{x}^2}{SS_{xx}})} = \sqrt{S_{\epsilon}^2(\frac{1}{n} + \frac{\bar{x}^2}{SS_{xx}})}$

$\bar{y} - b_1 \bar{x}$

$\hat{y} = a + b \cdot x$

$e^2 = (y - \hat{y})^2$

$\sum e^2 = \sum (y - \hat{y})^2 = SSE$

$MSE = \frac{SSE}{N-K}$

$SE_{b_1} = \sqrt{Var(b_1)} = \frac{MSE}{SS_{xx}}$

$t_{b_1} = \frac{b_1}{SE_{b_1}}$

$SE_{b_0} = \sqrt{Var(b_0)} = \sqrt{MSE(\frac{1}{n} + \frac{\bar{x}^2}{SS_{xx}})}$

$t_{b_0} = \frac{b_0}{SE_{b_0}}$

$SSM = SST - SSE$

Regression

[DataSet5]

Variables Entered/Removed^b

Model	Variables Entered	Variables Removed	Method
1	cost ^a	.	Enter

- a. All requested variables entered.
b. Dependent Variable: output

Model Summary

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate
1	.953 ^a	.909	.908	.34875

- a. Predictors: (Constant), cost

$SSE = SST - SSM$
 $MSE = SSE / (N - k)$ to be used in $\frac{MSM}{MSE}$, $Var(b)$, $Var(a)$
 $SEE = \sqrt{MSE}$
 standard error of residual
 $S_e = \sqrt{S_e^2}$
 $MSE = \text{Variance of residual}$
 $S_e^2 = \hat{\sigma}_e^2$

ANOVA^b

Model		Sum of Squares	df	Mean Square	F	<u>P-value</u> <u>Sig.</u>
1	Regression	SSM 107.123	K-1 1	MSM 107.123	880.730	.000 ^a
	Residual	SSE 10.703	N-k 88	MSE .122	$\frac{MSM}{MSE}$	
	Total	SST 117.827	N-1 89			

- a. Predictors: (Constant), cost **IV**
b. Dependent Variable: output **DV**

P-value < .05, Reject H₀ of $\beta_0 = \beta_1 = 0$

Coefficients^a

Model		Unstandardized Coefficients		Standardized Coefficients	t	<u>P-value</u> <u>Sig.</u>
		B	Std. Error	Beta		
1	(Constant)	-14.128	.438		-32.253	.000
	cost	.969	.033	.953	29.677	.000

- a. Dependent Variable: output

$a = \bar{y} - b\bar{x}$
 $b = \frac{SP_{xy}}{SS_{xx}}$
 $SE_{b_1} = \sqrt{\frac{MSE}{SS_{xx}}}$
 $SE_{b_0} = \sqrt{MSE \left(\frac{1}{n} + \frac{\bar{x}^2}{SS_{xx}} \right)}$
 $= \sqrt{S_e^2 \left(\frac{1}{n} + \frac{\bar{x}^2}{SS_{xx}} \right)}$
 $\frac{b_1}{SE_{b_1}} = \frac{.969}{.033}$
 $\frac{b_0}{SE_{b_0}} = \frac{-14.128}{.438}$
 $b_1 \frac{S_x}{S_y} = .969 * \frac{1.13199}{1.150606}$

"For a standard deviation increase in x, y is expected to change by $b_1 \frac{S_x}{S_y}$ standard deviations, holding all other variables constant."



This is not β_1 , but "standardized b_1 ."